<b>Motivation</b> ○	Dataset ⊙	Method 00		Conclusion
_				
	Which Airline is This?	Airline Logo D	Detection in Real-Wo	orld
		/eather Conditio		

<u>Christian Wilms</u>\*, Rafael Heid<sup>\*†</sup>, Mohammad Araf Sadeghi<sup>\*</sup>, Andreas Ribbrock<sup>†</sup>, and Simone Frintrop<sup>\*</sup>

 $^{\ast}$  University of Hamburg, Department of Informatics  $^{\dagger}$  zeroG GmbH, Germany





13 January 2021 ICPR 2020

Motivation •	Dataset ⊙	Method ○○	Conclusion
Characteristics of	Logo Detection		

## Logo Detection

- Used for checking visibility of advertisements, ...
- Mostly based on object detectors like Faster R-CNN
- Datasets feature clean images



Results of Faster R-CNN

Motivation •	Dataset ⊙	Method 00	Conclusion
Characteristics of	Logo Detection		

## Logo Detection

- Used for checking visibility of advertisements, ...
- Mostly based on object detectors like Faster R-CNN
- Datasets feature clean images
- Are object detectors enough?
- What about real-world weather?



Results of Faster R-CNN

Motivation •	Dataset ○	Method 00	Conclusion
Characteristics of	f Logo Detection		

## Logo Detection

- Used for checking visibility of advertisements, ...
- Mostly based on object detectors like Faster R-CNN
- Datasets feature clean images
- Are object detectors enough?
- What about real-world weather?





## Results of Faster R-CNN

- ... dataset with images captured in adverse weather conditions
- ullet ... system for airline logo detection + learning free data augmentation strategy

<b>Motivation</b> O	Dataset •	Method 00	Conclusion
Proposed D	ataset		

- Detection of airline logos on images from planespotters
- 7038 annotated airline logos 41 classes

	Train	Val	Test1	
Simple images	1	1	1	



Simple test split

Motivation	Dataset	Method	Conclusion
	•		
Proposed D	ataset		

- Detection of airline logos on images from planespotters
- 7038 annotated airline logos 41 classes

	Train	Val	Test1	Test2
Simple images	1	1	1	/
Difficult images				~

## Difficult images

- Only 252 images with 459 annotations
- Effected by adverse weather conditions



## Simple test split



## Difficult test split

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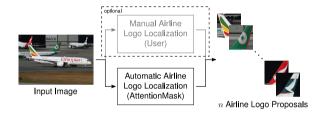
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<b>Motivation</b> O	Dataset ⊙	Method ●○	Conclusion
Airline Logo	Detection Syste	em	



Input Image

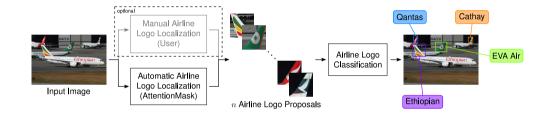
Motivation ○	Dataset ○	Method ●○	Conclusion
Airline Logo D	etection Svsten	n	



## Localization

- AttentionMask [Wilms and Frintrop, ACCV'18] for localizing logo proposals
- Can be provided by the user

<b>Motivation</b> O	Dataset O	Method ●○	Conclusion
Airline Logo	Detection Syste	m	



## Localization

- AttentionMask [Wilms and Frintrop, ACCV'18] for localizing logo proposals
- Can be provided by the user

#### Classification

- Classifier based on VGG
- Optimized architecture
- Lightweight design with 7 layers

Motivation	Dataset	Method	Conclusion
		00	

# Data Augmentation for Adverse Weather Conditions

Fog





## Original simple images





## Our augmentation on simple images





Increase brightness

Original difficult images

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Motivation	Dataset	Method	Conclusion
		00	

# Data Augmentation for Adverse Weather Conditions

Fog











Increase brightness

## Rain/dark clouds













Contrast reduction

Motivation	Dataset	Method	Conclusion
		00	

# Data Augmentation for Adverse Weather Conditions

Fog











Increase brightness

## Rain/dark clouds

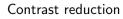












## Raindrops











Local blurring

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Motivation	Dataset	Method	Evaluation	Conclusion
○	○	00	•	
Results				

:		
mAP	mAP <sub>75</sub>	
0.698	0.869	
0.659	0.826	
0.708	0.880	
	mAP 0.698 0.659	mAPmAP <sub>75</sub> 0.6980.8690.6590.826



Faster R-CNN w/o DA



Ours w/o DA

Faster R-CNN [Ren et al., NIPS'15] YOLOv3 [Redmond and Farhadi, arXiv'18]

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<b>Motivation</b>	Dataset	Method	Evaluation	Conclusion
O	○	○○	•	
Results				

:		
mAP	mAP <sub>75</sub>	
0.698	0.869	
0.659	0.826	
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	mAP 0.698 0.659	mAPmAP <sub>75</sub> 0.6980.8690.6590.826



Faster R-CNN w/o DA



Ours w/o DA

Difficult Test Split					
mAP	mAP <sub>75</sub>				
0.101	0.129				
0.118	0.102				
0.173	0.221				
	mAP 0.101 0.118				

Faster R-CNN [Ren et al., NIPS'15] YOLOv3 [Redmond and Farhadi, arXiv'18]

Motivation	Dataset	Method	Evaluation	Conclusion
○	○	00	•	
Results				

Simple Test Split			
Method	mAP	mAP <sub>75</sub>	
YOLOv3	0.698	0.869	
Faster R-CNN	0.659	0.826	
Ours	0.708	0.880	





## Faster R-CNN w/o DA Ours w/o DA

## Difficult Test Split

Method	mAP	mAP <sub>75</sub>
YOLOv3	0.101	0.129
Faster R-CNN	0.118	0.102
Ours	0.173	0.221

#### Difficult Test Split with our Data Augmentation

Method	mAP	mAP <sub>75</sub>
YOLOv3	0.123	0.147
Faster R-CNN	0.128	0.117
Ours	0.203	0.248

Faster R-CNN [Ren et al., NIPS'15] YOLOv3 [Redmond and Farhadi, arXiv'18]

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Motivation	Dataset	Method	Conclusion
○	⊙	00	●○
Summarv			

- Dataset with adverse weather effects
- Tailored architecture for airline logo detection system
- Data augmentation strategy to counter adverse weather effects

## Conclusion

- Tailored architecture works better than object detectors
- Adverse weather conditions have strong effect on the results
- Data augmentation improves the results



Ours with DA

<b>Motivation</b>	Dataset	Method	Conclusion
O	O	00	○●
Thank you	for your attention	۱I	

#### Acknowledgement

We thank the administrators of planespotters.net for allowing us to use their database for this research.

# Visit our poster in session **PS T3.5** on **Wednesday**, 13 January

Los<br/>AngelesNew YorkCETBeijingSydney07:30 am10:30 am04:30 pm11:30 pm02:30 am